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**East European chironomid-based calibration model for past summer
temperature reconstructions**

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Running page head: East European chironomid-temperature model

26 ABSTRACT: Understanding local patterns and large scale processes in past climate necessitates
27 detailed network of temperature reconstructions. In this study, a merged temperature inference
28 model using fossil chironomid (Diptera: Chironomidae) datasets from Finland and Poland was
29 constructed to fill the lack of an applicable training set for East European sites. The developed
30 weighted averaging-partial least squares (WA-PLS) inference model showed favorable performance
31 statistics suggesting that the model can be useful for downcore reconstructions. The combined
32 calibration model includes 212 sites, 142 taxa and a temperature gradient of 11.3-20.1 °C. The 2-
33 component WA-PLS model has a cross-validated coefficient of determination of 0.88 and a root
34 mean squared prediction error of 0.88 °C. We tested the new East European temperature transfer
35 function in chironomid stratigraphies from a Finnish high-resolution short-core sediment record and
36 a Polish paleolake (Żabieniec) covering the past ~20,000 yr. In the Finnish site, the chironomid-
37 inferred temperatures correlated closely with the observed instrumental temperatures showing
38 improved accuracy compared to estimates by the original Finnish calibration model. In addition, the
39 long-core reconstruction from the Polish site showed logical results in its general trends compared
40 to existing knowledge on the past regional climate trends, however, with distinct differences when
41 compared with hemispheric climate oscillations. Hence, based on these findings, the new
42 temperature model will enable more detailed examination of long-term temperature variability in
43 Eastern Europe, and consequently reliable identification of local and regional climate variability of
44 the past.

45

46 KEYWORDS: Chironomidae, Climate reconstruction, Finland, Holocene, Late Glacial,
47 Paleoclimate, Poland, Training set, Transfer function

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50

51 1. INTRODUCTION

52

53 Advances in paleoclimatology have enabled building of comprehensive outline of climate
54 changes of the recent past, the Holocene epoch and the last Glacial cycle (McCarroll 2015, Wanner
55 et al. 2015, Linderholm et al. 2018). However, the local differences and small-scale variation are
56 still poorly established in several geographic areas. In addition to high-fidelity sediment archives,
57 the paleoclimatological toolpack needs to be refined to tackle reliably the climate variability of the
58 past. Non-biting midges (Insecta: Diptera: Chironomidae) have been recognized as one of the most
59 powerful proxies to reconstruct past summer air temperature dynamics (Brooks 2006). Utilizing the
60 fossil community compositions of the temperature-sensitive chironomid taxa and applying the
61 calibration set approach via a transfer function, quantitative climate inferences have become
62 available from sites where other paleoclimate proxies have failed or are not possible to use
63 (Ilyashuk et al. 2011, Luoto et al. 2018). In addition to confounding environmental variables, such
64 as nutrients (Quinlan & Brodersen 2006, Eggermont & Heiri 2012, Medeiros et al. 2015), a
65 potential downside of chironomids as a paleotemperature proxy lays in the suitability of the
66 calibration set to the downcore site (Engels et al. 2014). In an ideal situation, the downcore site
67 should be within the geographical area of the training set, the study site characteristics (such as lake
68 size and depth) should be similar and the calibration sites should constitute a temperature gradient
69 that covers the expected range of past temperature changes. When applying inference models to
70 cores outside the training set's geographical or environmental range, problems related to taxa
71 occurrences (poor modern analogues) and unrealistic taxon-specific temperature optima arises.
72 Moreover, continental scale calibration sets (Heiri et al. 2011) may not be able to detect small-
73 magnitude variation in temperatures, although they can be very useful in reconstructing the large-
74 scale climate patterns.

75 Previously, it has been challenging to produce reliable chironomid-based temperature
76 inferences at the ends of the temperature gradient in Eastern Europe. In downcore sites located in
77 southern Finland, temperatures of the warm climate events, such as the recent warming, Medieval
78 Climate Anomaly and Holocene Thermal Maximum, may have been underestimated due to lack of
79 equally warm calibration sites (Rantala et al. 2016, Shala et al. 2017). Similarly, the lack of warm
80 calibration sites in the available chironomid-based temperature inference models have thus far
81 caused problems in downcore studies of Polish sites due to deficiency of warm analogues
82 (Pawłowski et al. 2015, 2016a). Here, we combine the Finnish calibration sets (Luoto 2009, Luoto
83 et al. 2016) with a dataset collected from Poland (previously unpublished) to create a more
84 applicable temperature inference model for East European sites than has previously been available.
85 In addition to standard numerical testing of the model performance, we validate the model using a
86 chironomid stratigraphy from an annually laminated lake sediment record from Finland and
87 compare the reconstruction against instrumentally measured temperatures. In addition, we apply the
88 East European calibration model on a chironomid record from a Polish paleolake covering the past
89 ~20,000 yr and compare the output against previous reconstructions. We aim to produce a new
90 quantitative tool for more reliable reconstructions of past climate patterns in the East European
91 sector to better describe local climate variability.

92

93 **2. MATERIAL AND METHODS**

94

95 **2. 1. Study sites and sediments**

96

97 The training set study sites comprise of 212 lakes located in Finland and Poland (Fig. 1).
98 The 114 Finnish sites, collected with a Limnos gravity corer between 2005 and 2014, originate from
99 two previously published datasets located at a treeline transect in northeastern Lapland (32 lakes,

100 68°47' – 69°55'N) (Luoto et al. 2016) and along the latitudinal gradient of Finland (82 lakes, 60°13'
101 – 69°53'N) (Luoto 2009). The mean July air temperature in the Finnish sites varies between 11.3
102 and 17.1 °C (mean 14.4 °C, median 14.1 °C) within an altitudinal gradient of 4-405 m a.s.l. All the
103 sites are small and shallow (0.5-7.0 m) with pH between 4.6 and 8.4. The 98 Polish lakes, sampled
104 in summer 2014 using a Kajak corer, are located between 49°19'–54°68'N and constitute an
105 altitudinal gradient of 4-1624 m a.s.l. The mean July air temperature in the Polish sites varies
106 between 11.6 and 20.1 °C (mean 18.6 °C, median 18.9 °C), whereas the depth range is 0.3-15.0 m.
107 The lake water pH fluctuate between 5.1 and 9.8. The Polish dataset is previously unpublished.
108 Comparison between the combined training sets is given in Table 1.

109 The short-core test site Lake Nurmijärvi (61°35'N, 25°55'E; 87.7 m a.s.l.) is located in
110 south-central Finland (Fig. 1). The lake with annually laminated sediments is currently
111 circumneutral (pH = 7.0) and mesotrophic. The mean July air temperature at the study site is 16.9
112 °C (climate normals 1981–2010, Finnish Meteorological Institute). The sediment sequence was
113 cored in winter 2016 using a HTH-corer and subsampled at 1-cm intervals. The average sample
114 interval in the verified varve chronology (Ojala et al. 2016, 2018) is 4 years and the available
115 meteorological data begins from the 1830s. The full chironomid stratigraphy of Nurmijärvi is
116 published (Luoto & Ojala 2016).

117 The long-core sediment site Żabieniec (51°51'N; 19°46'E; 180 m a.s.l.) is currently a bog
118 located in central Poland (Fig. 1). The present-day mean July air temperature at the study site is 18
119 °C. Detailed descriptions of the study site and the paleolake sediments together with the full
120 chironomid stratigraphy and chronology are given elsewhere (Płóciennik et al. 2011). In brief, the
121 paleolake sediments were sampled using a piston corer and the subsampling was performed at
122 varying intervals. The stratigraphy represents roughly the past 20,000 yr. The chronology of the
123 core is based on 13 radiocarbon dates and the age-depth model is originally presented in
124 Lamentowicz et al. (2009).

125

126 2.2. Chironomid analysis

127

128 Fossil chironomid analysis was performed using standard methods in all the datasets and
129 cores applying provided guidelines (Brooks et al. 2007). In short, a 100 µm mesh was used for
130 sieving at least 50 chironomid head capsules per sample. Similar to unidentified remains, other
131 midges than chironomids were ignored. Taxonomic harmonization of the training sets and the two
132 sediment downcores was achieved through close collaboration between the chironomid analysts.
133 The morphologically similar taxa *Thienemanniola* and *Constempellina* were separated in the
134 combined training set according to their contemporary occurrence described in species checklists
135 (Paasivirta 2014, Sæther and Spies 2013). In some cases (including *Ablabesmyia*, *Dicrotendipes*
136 and *Microtendipes*), species type-level identification was scaled to genus-level.

137

138 2.3. Statistical analyses

139

140 Taxon-specific mean July temperature optima in the merged dataset were estimated using
141 Weighted Averaging (WA) with log10 transformed species data in the program C2 version 1.7.2
142 (Juggins 2007). Generalized Linear Modeling (GLM) was used to assess taxa that significantly ($p \leq$
143 0.05) respond to mean July air temperature. The GLMs were run using Poisson distribution in the
144 program Past3 (Hammer 2001). Detrended Correspondence Analysis (DCA) was used to assess the
145 gradient lengths of the first two DCA axis for selection of the most suitable methods for further
146 analyses. For linearly distributed data with short gradient lengths, Principal Component Analysis
147 (PCA) and redundancy analysis (RDA) are the most suitable methods (Šmilauer & Lepš 2014). The
148 primary PCA axis scores were compared with site-specific temperatures using Pearson Product-
149 moment correlation coefficient (R), coefficient of determination (R^2) and the level of statistical

150 significance ($p < 0.05$) to verify that the communities are responding to temperature. In addition,
151 RDA with forward-selected environmental variables and 999 unrestricted permutations was used to
152 partial out the significance of temperature, depth and pH (variables available from all datasets) on
153 the chironomid assemblages in the joint dataset. The DCA, PCA and RDA were performed with
154 log10 transformed species data using the CANOCO 5 program (Šmilauer & Lepš 2014).

155 The combined East European chironomid-based calibration model of mean July air
156 temperature was developed using the Weighted Averaging - Partial Least Squares technique (WA-
157 PLS), also with log10 transformed species data. The number of useful regression calibration
158 components was assessed using t -test (significance level 0.05). Model performance was evaluated
159 using jackknife cross-validation and subsequent coefficient of determination (R^2_{Jack}), root mean
160 squared error of prediction (RMSEP) and mean and maximum biases. The model was constructed
161 using the program C2 version 1.7.2 (Juggins 2007), in which also other common model types were
162 initially tested.

163 The model was verified against instrumentally measured (meteorological) temperatures
164 available since the 1830s in the shortcore sediment record from Nurmijärvi. The chironomid-
165 inferred temperatures were tested against the observational data by applying R , R^2 and $p < 0.05$.
166 Sample-specific modeling errors (estimated standard error of prediction = eSEP) were determined
167 using bootstrapping cross-validation with 999 iterations. The model was also run to reconstruct
168 temperatures in the Żabieniec long-core sediment record. LOESS smoothing was used to depict
169 general trends using a span of 0.2. To test whether the Żabieniec reconstruction corresponded to the
170 primary chironomid community variability, the temperatures were compared against the PCA axis 1
171 scores using Pearson product-moment correlation coefficient and the associated level of statistical
172 significance. Using the modern analogue technique, the cut-level of the 5th percentile of all squared-
173 chord distances in the modern calibration data was determined. These distances were then compared
174 to the distance between each fossil assemblage and its most similar assemblage in the modern data

175 set and used to define ‘no close’ analogues. The reconstruction was compared with previous
176 chironomid-based reconstructions from the focal core using the Norwegian (Brooks & Birks 2001,
177 unpubl.), Russian (Self et al. 2011) and Swiss (Heiri & Lotter 2005, Bigler et al. 2006, von Gunten
178 et al. 2008) calibration models. In addition, the reconstructed general local trends in temperature
179 were compared with an ice-core temperature record (site-specific calibrations using ice-isotopic
180 ratios, borehole temperatures and gas-isotopic ratios) from Greenland (GISP2, Cuffey & Clow
181 1997, Alley 2000) representing hemispheric climate development.

182

183 3. RESULTS

184

185 After merging the Finnish and Polish chironomid training sets, 142 taxa were encountered
186 from the 212 calibration sites (Fig. 2). *Psectrocladius sordidellus*-type occurred in 86 sites,
187 *Polypedilum nubeculosum*-type in 80 sites and *Dicrotendipes* and *Procladius* in 79 sites.
188 *Limnophyes* reached the maximum relative abundance (75%) in a single site. *Lauterborniella*
189 *agrayloides* (6.7%), *Ablabesmyia* (6.5%) and *Paratendipes nudisquama*-type (5.4%) had the
190 highest mean abundances in the combined dataset.

191 The taxa with coldest temperature optima (12.5-13.1 °C) included *Heterotrissocladius*
192 *maeeri*-type, *Psectrocladius calcaratus*-type and *Zalutschia* type B, whereas the taxa having
193 warmest optima included *Polypedilum sordens*-type, *Glyptotendipes barbipes*-type and *Labrundinia*
194 *longipalpis* (18.8-18.9 °C) (Fig. 3). Taxa with intermediate temperature optima (16-17 °C) and wide
195 tolerances included *Paratanytarsus penicillatus*-type, *Dicrotendipes*, *Procladius* and *Chironomus*
196 *anthracinus*-type. Of the most common taxa (N > 5), only *Dicrotendipes* did not respond
197 statistically significantly to the temperature gradient (Fig. 3). For most taxa, the GLMs showed
198 significant linear fit, however, significant nonlinear distribution was found in some taxa with
199 intermediate temperature optima, including *Tanytarsus chinyensis*-type 1, *Natarsia Punctata*-type,

200 *Corynoneura lobata*-type, *Smittia* and *Endochironomus impar*-type. *Paratanytarsus penicillatus*-
201 type was the only taxon with bimodal distribution, with highest abundances at the both ends of the
202 temperature gradient.

203 The initial DCA indicated a gradient length of 2.8 SD for the surface sediment chironomid
204 assemblages. Hence, owing to the linear nature of the data, PCA was recommended for ordination
205 analysis (Šmilauer and Lepš, 2014). Subsequently, the PCA axis 1 showed an eigenvalue of 0.1974
206 and the axis 2 an eigenvalue of 0.0659. The first axis explained 19.7% and the second 6.6% of the
207 total variance. The first four axes explained 36.1% of the variance in total. In the ordination (Fig. 4),
208 the samples along the primary PCA axis were arranged according to the site-specific mean July air
209 temperatures, with Polish sites (warm) having negative scores and Finnish sites (cold) positive
210 scores (Fig. 4a). The warm and cold indicator taxa identified with the PCA ordination (Fig. 4b)
211 were the same as indicated with the WA optima and GLMs (Fig. 3). The PCA axis 1 scores of the
212 samples were strongly correlated with the site-specific temperatures having an R of 0.91, R^2 of 0.82
213 and $p < 0.001$. The RDA results showed that temperature was the most important variable in
214 explaining chironomid distribution of the examined variables. Of variation explained by the
215 examined variables (15.4%), temperature explained 78.8%, pH 13.7% and depth 7.5% (Table 2).
216 Consequently, temperature had clearly the highest $\lambda_1:\lambda_2$ ratio (1.061) that justified the construction
217 of the chironomid-based temperature model.

218 Compared to other model types, WA-PLS had the best performance statistics with
219 respect to its R^2_{Jack} and RMSEP (Table 3). The developed WA-PLS model for mean July air
220 temperature had an R^2_{Jack} of 0.88, RMSEP of 0.88 °C and mean and maximum biases of -0.02 and
221 0.79 °C, respectively (Table 4). Addition of the second regression calibration component reduced
222 the RMSEP by 8.8% (randomization t -test significance 0.004). The 1:1 relationship between the
223 inferred and observed temperatures in the model illustrated that the combined calibration set has a

224 well-structured continuum in its temperature range with relatively even distribution of samples (Fig.
225 5a).

226 The test of the developed model on the clastic-biogenic varve record from Lake Nurmijärvi
227 showed similar trends between the chironomid-inferred and meteorologically observed
228 temperatures over the instrumental period. In both inferred and observed records (Fig. 6a), the
229 temperatures remained low during the 19th century with increased temperatures at the 1930s.
230 Following intermediate summer temperatures, the climate began to warm in the 1990s and record
231 highest temperatures synchronously occurred during the 21th century. The correlation between the
232 observed and inferred temperatures at the test site was statistically significant ($R = 0.72$, $R^2 = 0.52$,
233 $R_{\text{corrected}} = 0.51$, $p < 0.001$), although in several samples the temperature difference was larger than
234 the sample-specific error estimate (Fig. 6b).

235 In the long-core reconstruction, samples 1608-1181 cm (older than 15,000 cal yr BP) had
236 poor modern analogues according to the MAT suggesting that the early part of the sequence may
237 not be reliably reconstructed. Nonetheless, the reconstructed values correlated with the primary
238 PCA axis scores ($R=0.50$, $p<0.001$) indicating that chironomids do respond to the reconstructed
239 variable in the sediment profile. The chironomid-inferred temperature trends using the East
240 European model were rather similar to those reconstructed using the Norwegian, Russian and Swiss
241 models (Fig. 7). However, the new model reconstructed higher temperatures for the initial part of
242 the sediment record (1600-1500 cm, no age estimate), where poor modern analogues occurred. In
243 all, the East European model was most similar with the reconstruction derived using the Russian
244 model, whereas larger differences existed when compared with the results using the Norwegian and
245 Swiss models. In addition to the early part of the record, the East European model reconstructed
246 high temperatures between 16,000-12,000 cal yr BP and during the past ~1000 yr. Based on the
247 new model, the most distinct cold events occurred between 1400-1300 cm (ending at ~17,000-
248 16,000 cal yr BP) and 2000-1000 cal yr BP. However, the latter cold event occurs in samples with

low chironomid count sums and presence of semiterrestrial taxa (see Płóciennik et al. 2011 for details). When compared with the GISP2 record, the warm period at 15,000 cal yr BP and the following cooling is well represented. The rapid temperature rise during the early Holocene suggested by the GISP data and the late Holocene cooling trend in the current record suggest differences between the regional and global records.

4. DISCUSSION

4. 1. Training set

Combination of the Finnish and Polish chironomid datasets yielded a training set with a temperature gradient of 8.8 °C (11.3-20.1 °C) enabling wider usability with respect to paleoclimate reconstructions. The cold indicators (Figs 2, 3, 4b) in the combined dataset, such as *Heterotrissocladius maeaei*-type, *H. grimshawi*-type, *Psectrocladius calcaratus*-type, *Sergentia coracina*-type, *Micropsectra insignilobus*-type and *Tanytarsus lugens*-type are commonly found also in the cold lakes of other training sets from Eurasia (Heiri et al. 2011, Self et al. 2011). Similarly, the warm preferring chironomids, such as *Polypedilum sordens*-type, *Glyptotendipes barbipes*-type, *G. pallens*-type and *Endochironomus albipennis*-type are typical warm indicators in various other datasets (Heiri et al. 2003, Self et al. 2011). These warm taxa also appear to be more common in meso-eutrophic lakes, whereas the cold taxa are more often found in oligotrophic sites (Brooks et al. 2001, Brodersen & Quinlan 2006, Luoto 2011). This occurrence pattern has been previously documented (Eggermont & Heiri 2012) and is for large part related to the fact that warm lakes are often more productive and human influenced compared to the naturally oligotrophic cold lakes. In addition, the results showed that *Paratanytarsus penicillatus*-type, *Dicrotendipes*, *Procladius* and *Chironomus anthracinus*-type are eurythermic taxa having large temperature

274 tolerances (Figs 3, 4b) that has also been described in various other datasets (Larocque et al. 2006,
275 Fortin et al. 2015, Nazarova et al. 2015). These taxa aggregate several species that is probably one
276 of the reasons for their broad tolerance values. Of the most common taxa (Figs 2, 3), only
277 *Dicrotendipes* did not have a statistically significant relationship with temperature and *P*.
278 *penicillatus*-type was the only one having a bimodal distribution. These factors inevitably influence
279 their use as temperature indicators. Although the general temperature indication of the taxa is in
280 most part similar to the other training sets, there are significant differences in the values of the taxa-
281 specific temperature optima that are related to the temperature gradients of the respective datasets.
282 These regional differences in optima will become significant when selecting the training set to be
283 used in a downcore, seriously affecting the reconstructed quantitative values (Engels et al. 2014,
284 Fortin et al. 2015).

285 The PCA indicated a humped distribution of the samples in the ordination space (Fig. 4).
286 The samples were clearly distributed along the primary PCA axis according to their site-specific
287 temperatures that was also verified by the high correlation ($R = 0.91$, $R^2 = 0.82$, $p < 0.001$) between
288 the PCA 1 scores and observed temperatures at the sites. Although the ordination plot illustrates that
289 the secondary gradient also has influence on the assemblages, the PCA axis 2 explained only 6.6%
290 of the total variance. In addition, the RDA results (Table 2) clearly indicated that temperature is the
291 most significant variable explaining chironomid distribution among the mutually measured
292 variables in the Finnish and Polish datasets. Importantly, water depth, which has been found
293 significant in explaining intralake chironomid distributions in Finland (Luoto 2010) and elsewhere
294 (Kurek & Cwynar 2009, Engels et al. 2012, Luoto 2012a, 2012b), explained only a minor share of
295 the chironomid community compositions. Therefore, these results demonstrate that the chironomid
296 assemblages closely respond to mean July air temperature in the combined dataset, and
297 consequently justify the development of the East European chironomid-based temperature model.
298 The primary response of chironomids to temperature has been clearly evidenced in a bulk of

299 distributional studies (Larocque & Hall 2003, Nyman et al. 2005, Brooks et al. 2012) and their
300 biological response to temperature is also evident (Rossaro 1991, Eggermont & Heiri 2012).
301 Nonetheless, detecting the potential influence of secondary environmental gradients, such as water
302 depth, nutrients and DOC, on chironomid-based paleotemperature reconstructions remains
303 important, especially since their significance may vary in time (Nyman et al. 2008, Shala et al.
304 2014, Medeiros et al. 2015).

305

306 **4.2. Calibration model**

307

308 The developed temperature calibration model used WA-PLS technique, which
309 outperformed the other tested model types (Table 3). Typical for chironomid-temperature
310 calibration models (Heiri et al. 2011), the use of two WA-PLS components was statistically
311 justified. The model's statistical performance (Fig. 5), measured in R^2_{Jack} , was comparable with
312 other chironomid-based temperature models (Heiri et al. 2011, Holmes et al. 2011) but in its
313 RMSEP (0.88 °C, 10% of calibration set gradient) it outperformed several of the other models,
314 many of them having RMSEPs >1 °C (Barley et al. 2006, Porinchu et al. 2009, Nazarova et al.
315 2011). Compared to the new East European model, the original latitudinal Finnish temperature
316 model had lower R^2_{Jack} (0.78) but also lower RMSEP (0.72 °C, 12.4% of calibration set gradient)
317 (Table 4). However, since RMSEP is inherently influenced by the gradient length of the examined
318 variable, it may be more useful to compare the RMSEPs in relation to the temperature gradients. In
319 this sense, the RMSEP is more favorable in the combined model.

320 The combination of the Finnish and Polish datasets resulted as a consistent continuum in
321 the model's predictive abilities, as the Polish sites increased the temperature gradient of the model
322 towards warmer temperatures (Fig. 5a). Longer environmental gradients will help in situations
323 where past climate conditions approach the specific dataset's temperature limits (Birks et al. 2003).

324 Slight distortions at both ends of the temperature gradient were observed (Fig. 5b) that is inherent in
325 WA-PLS models (Heiri & Lotter 2010). This distortion, i.e. edge-effect, causes underestimation of
326 warm temperatures and overestimation of cold temperatures, and hence potentially smoothen
327 reconstructions.

328

329 **4.3. Reconstructions**

330

331 The best means to verify environmental reconstructions is to compare them with instrumentally
332 measured data (Larocque et al. 2009, Larocque-Tobler et al. 2015). Our test site, Lake Nurmijärvi
333 with sediments constituting of clastic biogenic varves, is located in southcentral Finland, close to
334 the warm end of the temperature gradient of the original Finnish training set (Fig. 1). The
335 reconstruction results showed that the East European model has the ability to accurately predict
336 downcore temperatures, as it well-depicted the cold temperatures of the 19th century, the increased
337 temperatures of the 1930s and the rapid warming that began in the 1980s (Fig. 6). The mean
338 difference in the inferred values compared to the observed was only 0.3 °C but the largest
339 overestimation was 3.0 °C and underestimation 2.7 °C. The biased values are most likely related to
340 lags in chironomid response times, since many taxa have long life cycles and their dispersal,
341 although fast compared to many other biological proxies (Wu et al. 2015), can take up to seven
342 years (Pinder 1986). The correlation between reconstructed and observed temperatures was higher
343 than in the previous study where the latitudinal Finnish model was used (Luoto & Ojala 2017),
344 clearly suggesting that the new model has better prediction accuracy and reliability compared to the
345 original model.

346 Since the model had solid performance statistics and it was able to reconstruct similar
347 temperatures with the observational record in southern Finland, we also applied it to a long-core
348 taken from the Żabieniec paleolake in Poland. Importantly, the chironomid-inferred values showed

349 significant correlation with the primary ordination axis scores verifying that chironomids respond to
350 temperature in the Żabieniec record. The previous study from the site demonstrated that
351 chironomid-based temperature models from outside the geographical area reconstructed partly
352 differing temperatures (Płóciennik et al. 2011). The present results showed that the developed East
353 European model reconstructed temperatures that mostly resemble those reconstructed using the
354 Russian model (Self et al. 2011) (Fig. 7), whereas distinct differences were apparent when
355 compared with the Swiss (Heiri & Lotter 2005) and Norwegian models (Brooks & Birks 2001)
356 (Fig. 7). These differences include very low temperatures in the initial part of the record and the
357 absence of the late Holocene cooling trend (Wanner et al. 2015). The early phase of the record may
358 be connected with the warm Kamion phase previously described from Poland (Manikowska 1995),
359 however, this remains uncertain due to lack of detailed chronological control in the bottom part of
360 the Żabieniec sediment sequence. Although the present interpretations are based solely on mean
361 July air temperature and there is no data on winter conditions or vegetational season length, it may
362 still be speculated that the climate conditions during the early phase of the record were glacial, but
363 because of high continentality summers were warm as in Siberia (Klimanov 1997). This
364 interpretation would be logical also considering that the East European and Russian models produce
365 similar warm temperatures for this phase differing from those derived using the Swiss and
366 Norwegian models (Fig. 6).

367 It is possible that the late Glacial temperatures could have been colder in Żabieniec than
368 the lowest temperatures represented by the calibration sites. However, the reconstructed
369 temperatures are not close the limits of the model (11 °C) but remain at >14 °C (Fig. 7). If the
370 actual temperatures would have been colder and the chironomid taxa would have consisted solely of
371 taxa with coldest temperature preferences, the WA-PLS model would have the potential to
372 extrapolate beyond the cold gradient end (Velle et al. 2011). This was not the case in the present
373 data, where late Glacial sequences consisted of taxa with intermediate temperature optima, such as

374 *Procladius* and *Tanytarsus pallidicornis*-type (Fig. 7). These taxa are also known to have wider
375 trophic tolerances (Brodersen & Quinlan 2006), which could reflect elevated nutrient condition
376 during the early part of the sediment profile. In contrast, the Norwegian, Russian and Swiss
377 calibration models all reconstruct consistently colder late Glacial temperatures than the East
378 European model. This is probably related to the compared calibration models having generally
379 colder lakes among the training set sites, and hence, the new model would benefit from inclusion on
380 even colder sites than it currently has. Despite the extrapolation capabilities of the WA-PLS
381 method, it is clear that the East European model can be reliable only within its temperature gradient
382 (11-20 °C), with decreasing reliability towards the gradient ends. Consequently, the coldest and
383 warmest episodes in long sediment records, such as the Żabieniec record, should be considered
384 cautiously when observing the reconstructed values, although at the same time the trends may be
385 realistic. It is also noteworthy that the late Glacial chironomid assemblages had poor modern
386 analogues in the calibration set that decreases the reliability of the new reconstruction during this
387 early phase of the record.

388 Similar to the other chironomid-based reconstructions, the new reconstruction did not
389 depict significant temperature drop during the cold Younger Dryas period. There are also no distinct
390 changes in the taxonomic composition at this time (Płóciennik et al. 2011, Fig. 7) that could
391 indicate other driving factors that would potentially reduce the temperature signal. This period was
392 unusually cold in Scandinavia (Brooks & Birks 2000, Wohlfarth et al. 2018) but in several studies
393 from Poland (Zawiska et al. 2015, Pawłowski et al. 2015, 2016a, 2016b) and Central Europe
394 (Larocque-Tobler et al. 2010), the summer temperature drop during the Younger Dryas has been
395 relatively muted compared to the British Isles, the Baltic region and the northern parts of the
396 continent (Heiri et al. 2014). Therefore, the present results are consistent with the previous studies
397 from Poland showing intra-European differences. Compared to the other chironomid-based
398 reconstructions from Żabieniec, the East European model reconstructs similar 16-17 °C

399 temperatures, with the exception of the Norwegian model, which indicates slightly lower
400 temperatures (Fig. 7).

401 Compared to hemispheric temperatures reflected by the GISP2 ice core record (Cuffey &
402 Clow 1997; Alley 2000), the current reconstruction does not suggest similar increase in early
403 Holocene temperatures (Fig. 7). The rapid early Holocene temperature increase has been described
404 from several lake sediment records from northern Europe (Brooks & Birks 2000, Engels et al. 2014,
405 Luoto et al. 2014, Shala et al. 2017, Helmens et al. 2018) and the European Alps (Samartin et al.
406 2012). The increase in *Lauterborniella* between 10,000 and 8000 cal yr BP can be related to
407 nutrient conditions, since in addition to high temperature optimum, it is known to thrive in more
408 nutrient-enriched lakes (Brooks et al. 2007). However, the warming associated with the increase in
409 *Lauterborniella* is consistent with the increased temperatures in the GISP2 record following the
410 cold early Holocene. The late Holocene cooling trend is not apparent in the GISP2 record, although
411 clearly seen from the present results and several other records from Poland (Zawiska et al. 2015,
412 Pawłowski et al. 2015, 2016b) suggesting regional deviation from hemispheric temperatures.
413 Nonetheless, it should be noted that the temperature decrease reconstructed from Żabieniec between
414 2000 and 1000 cal yr BP is not reliable owing to dominance of semiterrestrial chironomid taxa and
415 low count sums (Płóciennik et al. 2011). Compared to the reconstructions performed using the other
416 chironomid-based models and the GISP2 record, the temperatures reconstructed using the East
417 European model showed a distinct warming during the past 1000 yr, with a short-lived drop in
418 temperatures during the Little Ice Age. In general, the reconstruction of the Holocene temperatures
419 in Żabieniec paleolake closely resembles those from elsewhere in Europe (Davis et al. 2003, Luoto
420 et al. 2010, Engels et al. 2014) combined with distinct local features (Zawiska et al. 2015,
421 Pawłowski et al. 2015, 2016b), hence signifying the reliability of the reconstruction.

422

423 5. CONCLUSIONS

424

425 Merging the Finnish and Polish chironomid-based training sets resulted as a valid
426 calibration model for mean July air temperature with an extended temperature gradient and
427 improved applicability. The temperature indicators were similar to what has been found in previous
428 studies, however, the numerical optima more accurately adjusted for the study area. The statistical
429 tests showed that chironomids were responding most strongly to temperature, hence enabling
430 construction of the enhanced model. Compared to previous chironomid-temperature models in
431 general, the new East European model has solid performance statistics.

432 The model validation in a Finnish annually laminated lake sediment record, covering the
433 observational temperature period beginning from the 1830s, showed that the model better predicts
434 paleotemperatures compared to the original Finnish model. Since the inferred temperatures
435 correlated strongly with the instrumental record, the model can be considered solid with respect to
436 its predictive abilities and applicability in downcore profiles from Eastern Europe. The
437 reconstructed temperatures using the East European model in the long-core from the Polish
438 paleolake Żabieniec, covering the past ~20,000 yr, were more similar to temperatures reconstructed
439 using the Russian chironomid-based model than the ones reconstructed using the Swiss and
440 Norwegian models. The reconstructed temperature trends were comparable to previous studies from
441 Poland but significantly different from hemispheric paleotemperature estimates signifying the
442 importance of local reconstructions in understanding past climate oscillations.

443 Although the model is designed for Finnish and Polish sites as a preset, it can be useful in
444 other areas of Eastern Europe as well. In the future, the East European model can be further
445 developed especially by including additional calibration sites from the Baltic Countries that would
446 promote stability of the model. In all, the present results demonstrate the usability and sensitivity of
447 fossil chironomids as quantitative paleoclimate indicators.

448

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718 **Tables**

719 **Table 1.** Characteristics of the study sites in the Finnish, Polish and the combined East European
720 chironomid-based temperature datasets. Mean values are given in brackets.

	Finnish	Polish	Combined
Number of sites (N)	114	98	212
Number of taxa (N)	111	100	142
Latitude (°N)	60.13-69.55 (65.91)	49.19-54.68 (52.60)	49.19-69.55 (59.60)
Longitude (°E)	22.00-30.13 (26.55)	14.51-23.42 (18.42)	14.51-30.13 (22.69)
Elevation (m a.s.l.)	4-405 (157)	4-1624 (196)	4-1624 (174)
Temperature gradient (°C)	11.3-17.1 (14.4)	11.6-20.1 (18.6)	11.3-20.1 (16.4)
Sampling depth (m)	0.5-7.0 (2.3)	0.3-15.0 (8.8)	0.3-15.0 (5.7)
pH	4.6-8.4 (6.5)	5.1-9.8 (8.3)	4.6-9.8 (7.5)

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736 Table 2. Redundancy analysis (RDA) results for the combined East European chironomid dataset.

737 All examined variables together explain 15.4% of the total variance.

Variable	$\lambda_1:\lambda_2$	Contribution (%)	F	p ($p_{\text{bonferroni adjusted}}$)
Mean July air temperature	1.061	78.8	26.0	0.001 (0.003)
pH	0.563	13.7	4.6	0.001 (0.003)
Water depth	0.398	2.5	2.5	0.002 (0.006)

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757 Table 3. Comparison of performance statistics using different model types (WA = weighted
758 averaging, PLS = partial least squares) in the development of the East European chironomid-based
759 calibration model. The model deemed most suitable for downcore reconstructions is marked with
760 boldtype.

Calibration model type	Coefficient of determination (R^2_{jack})	Root mean squared error of prediction (RMSEP, °C)	Maximum bias (°C)	Reduction in RMSEP (%)
WA _{inverse deshrinking}	0.86	0.96	1.22	
WA _{classical deshrinking}	0.86	1.01	0.76	
PLS _{component 1}	0.84	1.04	1.51	
PLS _{component2}	0.86	0.96	0.93	7.99
WA-PLS _{component1}	0.86	0.97	1.24	
WA-PLS_{component2}	0.88	0.88	0.79	8.78

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775 Table 4. Performance statistics of the developed East European chironomid-based temperature
 776 calibration model compared with the original Finnish model (Luoto 2009).

	East European model	Finnish model
Number of sites (N)	212	82
Number of taxa (N)	142	110
Model type	WA-PLS, component 2	WA-PLS, component 2
Coefficient of determination (R^2_{jack})	0.88	0.78
Root mean squared error of prediction (RMSEP)	0.88 °C	0.72 °C
Maximum bias	0.79 °C	0.79 °C

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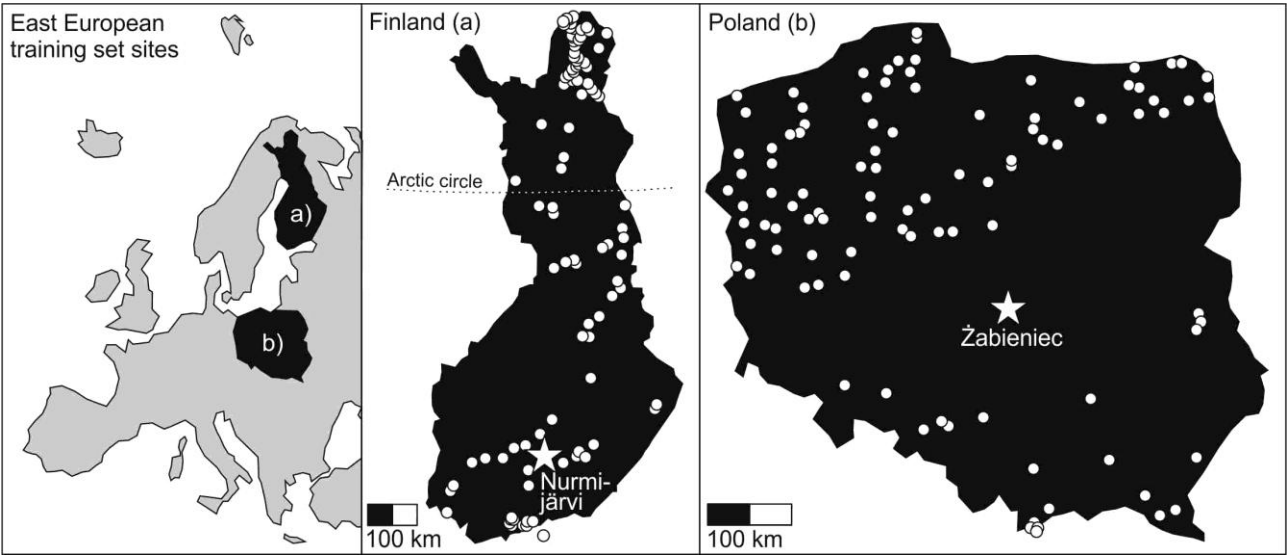
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795 **FIGURES**

796



797 Fig. 1. Calibration sites of the East European chironomid-based temperature training set. The
798 Finnish dataset ($60^{\circ}13' - 69^{\circ}55'N$) include 114 (a) and the Polish dataset ($49^{\circ}19' - 54^{\circ}68'N$) 98 lakes
799 (b). The downcore study site Żabieniec (Poland) and Nurmijärvi (Finland) are marked with stars.

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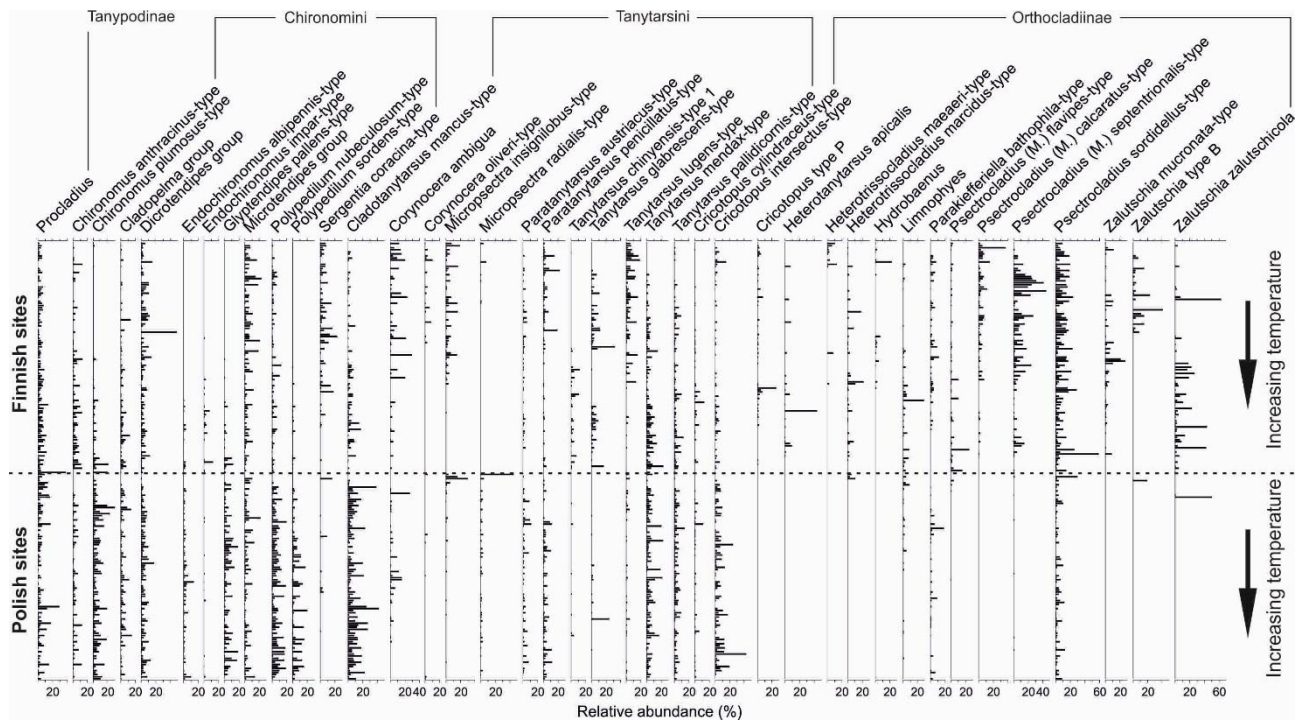
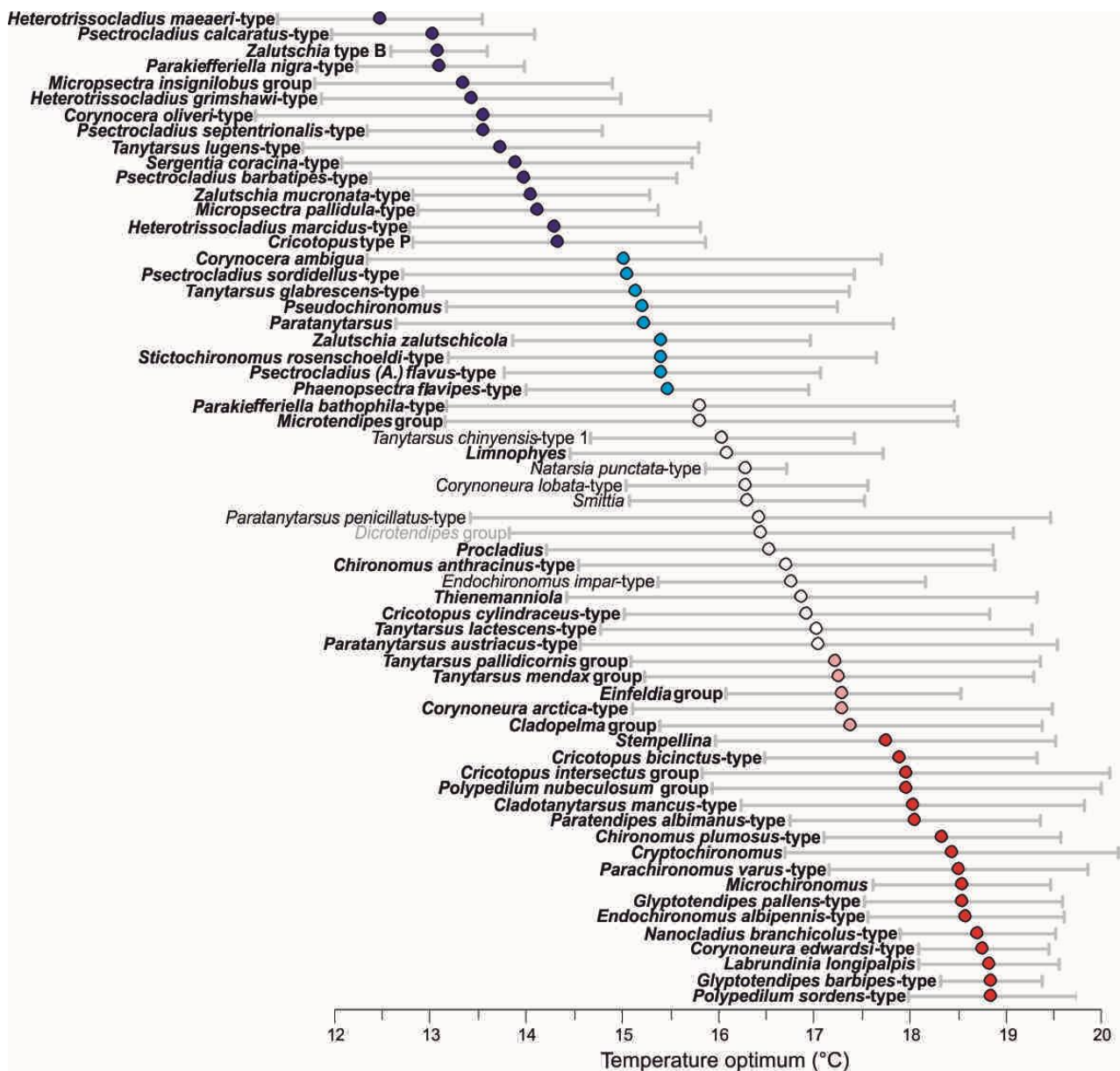


Fig. 2. Most common chironomid taxa (>10 occurrences and maximum abundance >10%) in the combined East European chironomid-temperature training set.



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817 Fig. 3. Chironomid mean July air temperature optima (weighted averaging) of the most common
 818 taxa (N > 5) in the combined East European dataset. The cold indicators are marked with blue,
 819 intermediate taxa with white and warm indicators with red. Taxa having statistically significant
 820 linear relationships with temperature are marked in bold type and nonlinear fit with regular font.
 821 The only taxon (*Dicrotendipes*) with no significant relationship is marked in grey.

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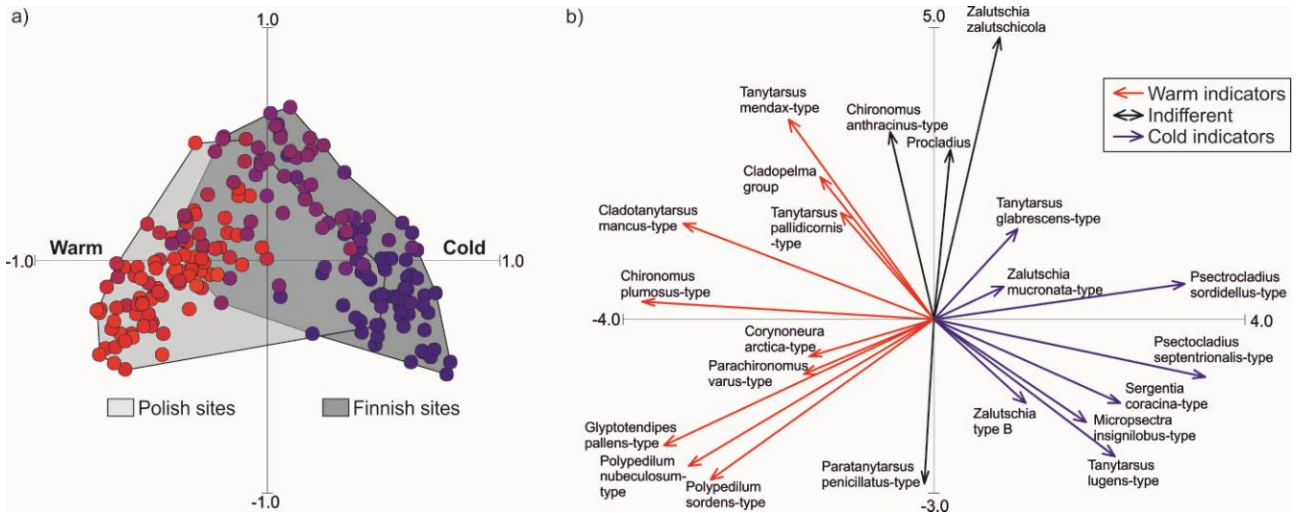


Fig. 4. Principal Component Analysis (PCA) ordination plots for samples (a) and selected taxa (b) based on surface sediment chironomid assemblages from lakes in Poland and Finland. The first (horizontal) PCA axis ($\lambda = 0.20$) explains 19.7% and the second (vertical) PCA axis ($\lambda = 0.07$) 6.6% of the total variance. The samples (sites) are colored according to their temperature from warm (red) to cold (blue) and the envelopes represent the different geographical regions.

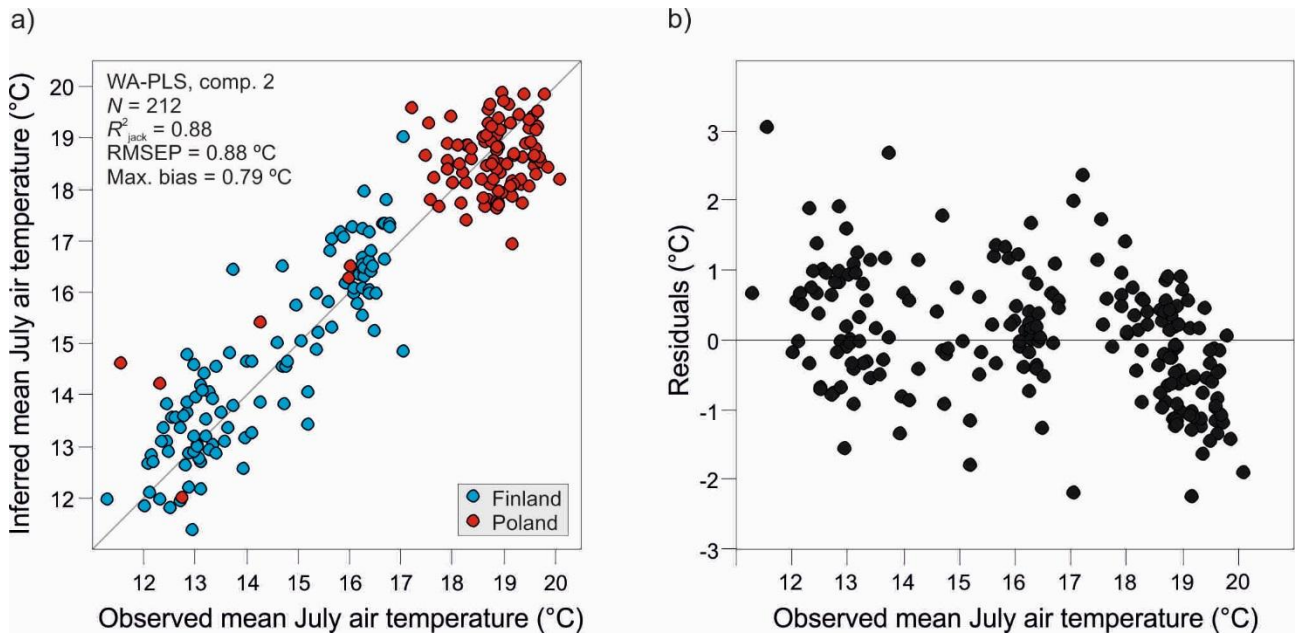
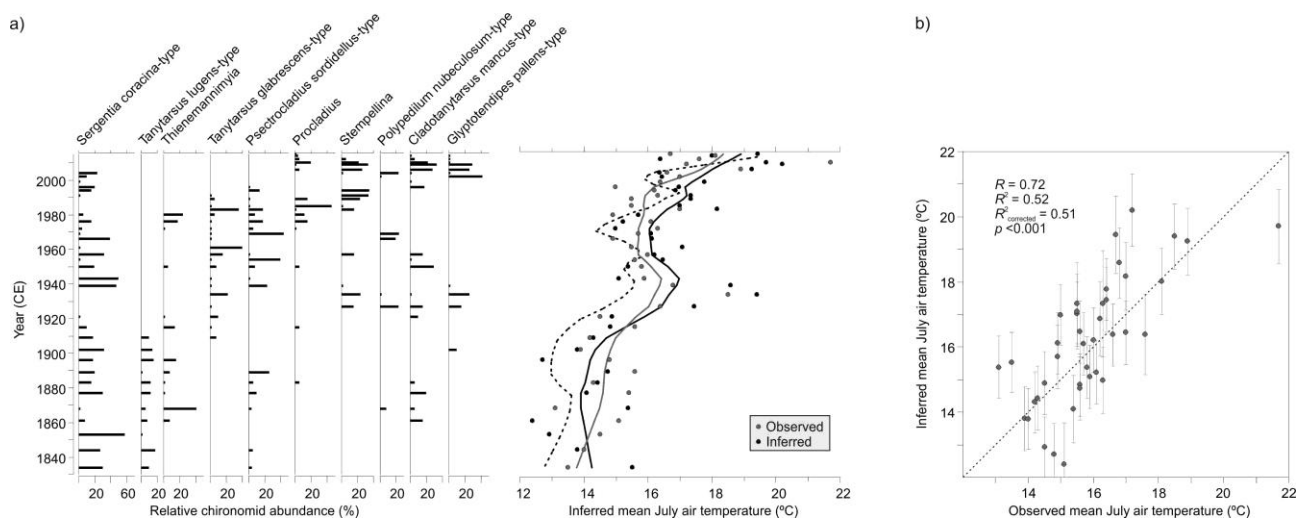


Fig. 5. (a) Relationship (1:1) between observed and chironomid-inferred mean July air temperatures in the East European calibration model using the Weighted Averaging-Partial Least Squares (WA-PLS) technique with two regression calibration components. N = number of calibration sites, R^2_{jack} = jackknife cross-validated correlation coefficient, RMSEP = root mean squared error of prediction. (b) Residuals versus observed temperatures.



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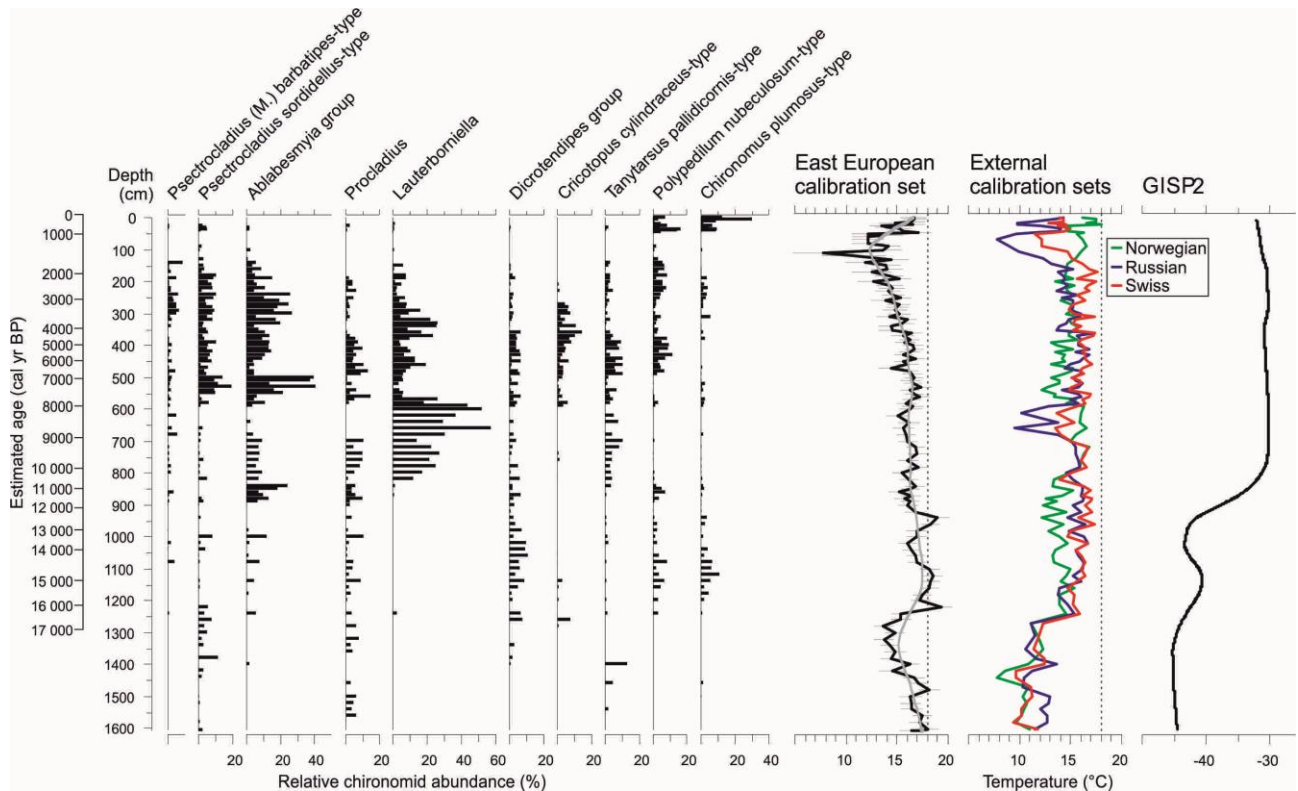
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Fig. 6. Ten most common chironomids (Luoto & Ojala 2017) and chironomid-inferred mean July air temperature reconstruction from Lake Nurmijärvi (southern Finland) using the East European calibration model compared with instrumentally measured temperatures (a) and their 1:1 relationship with sample-specific error estimates using bootstrapping cross-validation (b). The dashed curve is the smoothed (LOESS span 0.2) original reconstruction using the Finnish model (Luoto & Ojala 2017). The taxa are ordered according to their temperature optima in the calibration set from the coldest to warmest.



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Fig. 7. Ten most common chironomids (Płóciennik et al. 2011) and chironomid-inferred mean July air temperature reconstruction from Żabieniec paleolake (Poland) using the East European calibration model compared with reconstructions (Płóciennik et al. 2011) using the Norwegian, Swiss and Russian calibration datasets. The taxa are ordered according to their temperature optima in the calibration set from the coldest to warmest. The gray line in the new reconstruction represent LOESS smoothing (span 0.2) and the error bars represent the bootstrap estimated sample-specific errors. Modern temperature at the study site is drawn as dashed lines. The smoothed (0.2) Greenland ice core data (GISP2, Cuffey & Clow 1997; Alley 2000) is provided to illustrate hemispheric temperature development (note that the timescale is only tentative due to chronological uncertainties).